

Effects of imperfect quality items in the asymmetric information structure in supply chain model

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ABSTRACT

Most of the supply chain models have been developed under symmetric information structure i.e. players have complete knowledge of each other's policies. But in most of the cases, players do not have complete information about the other players i.e. some information regarding their businesses is hidden. This paper studies supply chain model of imperfect quality items under asymmetric information in which unit price taken by the buyer and unit marketing expenditure are influencing product's demand. This information is hidden to seller. The seller delivers the supply to the buyer. Each delivered lot goes through an inspection process at the buyer's side. In the inspection process, the items are divided into two categories. The first category is perfect quality items while the second category is of imperfect quality items. After inspecting the lots, the perfect quality commodities are to be sold at selling price and the imperfect items are supposed to get sold at a discounted price. The interaction between two constituents of the supply chain is modeled by non-cooperative Seller Stackelberg game approach in which the role of leader is played by the seller while the role of follower is played by the buyer. In our proposed model, the seller does not have information related to market demand but the buyer does. Numerical examples and sensitivity analysis explain the significance of the theory.

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1. Introduction

Game theory is the study and analysis of mathematical models of cooperation and human conflict in the presence of a competition. For some contexts, game theory is about the science of strategies, or at least the optimized decision-taking approach of independent and competitive actors in a strategic setting. The key role in game theory was played by mathematicians John Nash and John von Neumann, as well as economist Oskar Morgenstern. Various researches have been published related to supply-chain management. Sarmah et al. (2006) developed a buyer seller coordination models in supply chain channel. Different mechanisms have been brought related to supply chain like credit option, quantity discount (Chiang et al., 1994; Weng, 1995), buy back and quantity flexibility. Cachon and Lariviere (2005) explained the contract of sharing the revenue and compared this contract to various other supply chain contracts: sales-rebate contracts buy-back contracts, quantity-flexibility contracts, price-discount contracts, franchise contracts, and quantity discounts.

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Under the fixed demand, several researches (Sucky, 2005a, 2006b; Heuvel et al., 2007; Chan & Kingsman, 2007; Dai & Qi, 2007; Papachristos & Konstantaras, 2006) have contributed in the field of supply chain in which they determined the optimal lot size and order cycle to achieve maximum profit of supply chain channel. Initially, Schwaller (1988) explored an *EOQ* model on defective items under inspection cost. Salameh and Jaber (2000) extended the *E.O.Q./E.P.Q.* model for imperfect items and these imperfect items are to be sold at some discounted prices in the completion of inspection process in a single lot. By adding some constant parameter, Cárdenas-Barrón (2000) corrected the optimum order quantity formula obtained by Salameh and Jaber (2000). Goyal and Cárdenas-Barrón (2002) presented a simple way for EOQ model with items of imperfect quality and compared the result with Salameh and Jaber (2000) and obtained optimal results. Maddah and Jaber (2008) extended their work by analyzing the effect of screening speed and supply process' variability on the lot size. They illustrated in their model that when the divergence of the yield rate is lower, then the order quantity is higher as compared to the one in classical *EOQ*.

Fixed demand is avoided by many researchers in supply chain channel in which the optimal size of the lot and optimal price are determined to maximize its profit. Researchers like Lee (1993), Abad (1994), Lee et al. (1996), Kim and Lee (1998), Jung and Klein (2001a, 2005b) showed their interest to establish supply chain model to calculate the optimal policy for buyer and seller, in which final demand is price sensitive. With the same demand function Jaggi et al. (2011) constructed a model which maximizes the retailer's total profit and jointly optimize the price and number of replenishments.

Some another researchers (e.g. Freeland, 1980; Lee and Kim, 1993; Sadjadi et al., 2005) proposed models in which demand depends not only on selling price but also on marketing expenditure. Abad (1994) work was extended by Abad and Jaggi (2003), where seller presents credit to the buyer when the demand is price sensitive. In this case, seller and buyer interact with each other and their profits are maximized with optimal policies under non-cooperative and cooperative game theoretic approach. Esmaeli et al. (2009) also established cooperative and non-cooperative game supply-chain model in which demand depends on marketing expenditure and price.

In the above papers, buyer and seller have all the information about each other's activities. In many situation, some private information related to the business is hidden to other players. Harsanyi (1967, 1968a,b) developed the theory about analysis of highly innovative games of incomplete information called Bayesian games. Corbett and Groote (2000) derived the optimal quantity discount policy using asymmetric information and compared the results with the conditions, where the seller has full information. Lau and Lau (2005) presented an illustration of the introduction about information-asymmetry and stochasticity on 2-echelon supply chain gaming models and demonstrated its necessity and practicality of using it rather than the usual deterministic symmetric information way in various supply-chain models.

Lau et al. (2007) demonstrated a procedure using stochastic and asymmetric demand information for a superior manufacturer to create a profit-maximizing volume-discount scheme. Sucky (2006) presented a bargaining model regarding the buyer-cost structure considering that the buyer possesses the authority to impose its individual optimal policy under asymmetric information. Reyes (2005) extended the newsboy problem for buyer and seller in the asymmetric market information structure in which market demand is familiar to the buyer and presented the model in which efficiency of supply chain channel is improved by taking the advantage of the buyer by the seller. Esmaeli and Zeephongsekul (2010) extended his work on asymmetric-information game by using the same demand function. Zhang and Panlop (2013) proposed a supply chain model with credit option under symmetric as well as asymmetric information and also draw some conclusion by taking different probability density function.

Every production system may produce some defective items. Jaggi et al. (2013) suggested that defectives can be detected with the help of inspection process after the delivery which can impact the ordered quantity. Thus, inspection of the items cannot be ignored at the buyer's end. None of the authors worked on the imperfect quality items in supply chain models with symmetric and asymmetric information structure. In this paper, we assume that the rate of production of the seller is linearly

associated to demand rate. Also, the demand is sensitive to marketing expenditure and selling price. The seller and buyer interact with each other and this interaction is shown by several supply chain models with imperfect items under non-cooperative game with both asymmetric and symmetric information structure. Seller Stackelberg is illustrated under non-cooperative approach. In this approach, seller plays as leader and the buyer plays as follower. Numerical examples show that profit function of the symmetric case is more in comparison to the profit function obtained in the asymmetric case. We assume the seller's belief on the buyer's type follow uniform distribution. First section consists of Introduction and Literature review. Second section introduces notation and assumptions used in the paper. In third section, we summarize buyer's model, seller's model and primary concepts of asymmetric information game. Fourth section deals with the non-cooperative seller Stackelberg game in both asymmetric and symmetric information scenario. In fifth and six sections, numerical examples and sensitivity analysis are given. In the last two sections, observations and conclusion are provided.

2. Notations and assumptions

2.1 Notations

The notations used throughout the paper.

Decision variables

c_b Seller's selling price (decision variable of seller) (\$/unit)

Q Order quantity determined by the seller (decision variable of seller) (units)

M Marketing expenditure cost (decision variable of buyer) (\$/unit)

p_b Buyer's retail price (decision variable of buyer) (\$/unit)

Parameters

A_b Buyer's order placement cost (\$/order)

A_s Ordering cost of seller (\$/order).

H_b Inventory carrying cost

C Seller's unit purchasing cost (\$/unit)

I Percentage of inventory holding cost per unit per year.

T_1 Cycle time in years for the buyer, $T_1 = Q(1 - \alpha)/D$

T_2 Cycle time in years for the seller, $T_2 = Q/D$

T Cycle time in years for the seller Stackelberg model, $T = \text{Max}(T_1, T_2)$

α Percentage of defective items delivered to buyer by the seller

λ Buyer's screening rate (unit/year)

c_s Defective quality items' cost per unit (\$/year)

t Required time to screen the defective items, $t = Q/\lambda$ (Years)

k Demand function's scaling constant ($k > 0$)

r Production rate(units/day)

u Production function's scaling constant ($u \geq 1$)

e Demand function's price elasticity ($e > 1$)

β Marketing expenditure elasticity of demand function ($0 < \beta < 1, \beta + 1 < e$)

D Annual demand rate (unit/year), Demand is presumed as function of selling price p_b and marketing expenditure M i.e. $D = kp_b^{-e}M^\beta$

2.2 Assumptions

1. The annual demand is sensitive to marketing expenditure and selling price. (Esmaeili et al., 2009), which is known to the buyer through the parameters e, β but unknown to the seller.
2. Planning horizon is infinite.
3. Every lot has α percentage of defectives items of known probability density function.
4. Shortages are not permitted. We assume the production rate r and demand rate d is linearly related defined as $r = ud$, $u \geq 1$ (Esmaeili et al., 2009)
5. Generally, in supply chain model, ordered quantity is determined by the buyer but we assume order quantity or lot size is the decision variable of the seller.

3. Mathematical models

3.1 Buyer's model

The objective of the buyer is to find the retailer price, p_b , and marketing expenditure, M , such that the net profit gets maximized. In this paper, the demand relies on the marketing expenditure and selling price. We consider a supply chain model with imperfect quality items. The seller supplies the items to the buyer with unit price. After delivery of Q units (assumed), the lot goes through an inspection process at the buyer's end. Defective items and non-defective items are separated through the inspection process from the entire delivered quantity Q at the rate of λ units per unit time. We assumed that, in a lot of Q quantity items, α percent items are found defectives with known probability density function. Defective items αQ are sold in a single lot at the end of screening time, $t = Q/\lambda$ at a discounted price c_s and non-defective items, $(1-\alpha)Q$ are sold at a price p_b . Thus the total revenue of the buyer is $p_b(1 - \alpha)Q + c_s\alpha Q$ in which defective items αQ are sold at a discounted price c_s and non-defective items, $(1-\alpha)Q$ are sold at a selling price p_b . The purchasing cost of the buyer of Q quantity at a price c_b is $c_b Q$. The marketing cost of the buyer of Q quantity is MQ and ordering cost of the buyer is A_b , and inventory carrying cost will be equal to $H_b (\frac{Q(1-\alpha)T}{2} + \frac{\alpha Q^2}{\lambda})$, where $H_b = Ic_b$.

Total annual profit for the buyer is denoted by $TP_b(p_b, M)$ and is given by

$$TP_b(p_b, M) = \text{Sales Revenue} - \text{Purchasing cost} - \text{Marketing cost} - \text{Ordering cost} - \text{Inventory carrying cost} \\ = p_b(1 - \alpha)Q + c_s\alpha Q - c_b Q - MQ - A_b - \left(\frac{Q(1-\alpha)T_1}{2} + \frac{\alpha Q^2}{\lambda}\right) H_b$$

Put $T_1 = \frac{(1-\alpha)Q}{D}$, $t = \frac{Q}{\lambda}$, $H_b = Ic_b$ then buyer's profit becomes

$$TP_b(p_b, M) = p_b(1 - \alpha)Q + c_s\alpha Q - c_b Q - MQ - A_b - \left(\frac{Q^2(1 - \alpha)^2}{2D} + \frac{\alpha Q^2}{\lambda}\right) Ic_b \\ = p_b(1 - \alpha)Q + c_s\alpha Q - c_b Q - MQ - A_b - \left(\frac{Q^2(1 - \alpha)^2}{2D} + \frac{\alpha Q^2}{\lambda}\right) Ic_b$$

Thus, the total expected buyer's profit is given by

$$E[TP_b(p_b, M)] = p_b E[1 - \alpha]Q + c_s E[\alpha]Q - c_b Q - MQ - A_b - \left(\frac{Q^2(1 - \alpha)^2}{2D} + \frac{\alpha Q^2}{\lambda}\right) Ic_b \\ = p_b(1 - E[\alpha])Q + c_s E[\alpha]Q - c_b Q - MQ - A_b - \left(\frac{Q^2 E[(1 - \alpha)^2]}{2D} + \frac{E[\alpha]Q^2}{\lambda}\right) Ic_b$$

By using, Renewal theory as used in Maddah and Jaber (2008), we have the expected total profit per cycle for the buyer,

$$\begin{aligned} E[TP_{b,M}] &= \frac{E[TP_{b,M}]}{E[T_1]} \frac{D}{Q(1-E[\alpha])} E[TP_b(p_b, Q)] \\ &= \frac{D}{Q(1-E[\alpha])} \left[p_b(1 - E[\alpha])Q + c_s E[\alpha]Q - c_b Q - MQ - A_b - \left(\frac{Q^2 E[(1-\alpha)^2]}{2D} + \frac{E[\alpha]Q^2}{\lambda} \right) Ic_b \right] \\ &= p_b D + \frac{1}{(1-E[\alpha])} \left[c_s E[\alpha]D - c_b D - MD - \frac{A_b D}{Q} - \left(\frac{QE[(1-\alpha)^2]}{2} + \frac{E[\alpha]QD}{\lambda} \right) Ic_b \right] \end{aligned}$$

In this paper, D , demand function assumed as, $D = Kp_b^{-e}M^\beta$.

$$E[TP_{b,M}] = Kp_b^{-e+1}M^\beta + \frac{Kp_b^{-e}M^\beta}{(1-E[\alpha])} \left[c_s E[\alpha] - c_b - M - \frac{A_b}{Q} - \frac{E[\alpha]Q}{\lambda} Ic_b \right] - \left(\frac{QE[(1-\alpha)^2]}{2(1-E[\alpha])} Ic_b \right) \quad (1)$$

The aim is to find the optimal value of p_b and M which will maximize the expected profit of buyer. The necessary and sufficient condition for expected total profit per cycle must be satisfied. The optimal value of p_b can be obtained by differentiating Eq. (1) with respect to p_b for fixed M and setting it to zero,

$$\frac{\partial E[TP_{b,M}]}{\partial p_b} = 0, \text{ yields,}$$

$$p_b = \frac{e}{(e-1)(1-E[\alpha])} \left[M + c_b + \frac{A_b}{Q} + \frac{Ic_b E[\alpha]Q}{\lambda} - c_s E[\alpha] \right] \quad (2)$$

Given expected profit function is strictly pseudo concave (Bazaraa et al., 2013) with respect to p_b for fixed M substituting the value of Eq. (2) into Eq. (1), we get,

$$\begin{aligned} E[TP_{b,M}] &= \frac{K}{e} \left[\frac{e}{(e-1)(1-E[\alpha])} \left[M + c_b + \frac{A_b}{Q} + \frac{Ic_b E[\alpha]Q}{\lambda} - c_s E[\alpha] \right] \right]^{-e+1} M^\beta \\ &\quad - \left(\frac{QE(1-\alpha)^2}{2(1-E[\alpha])} Ic_b \right) \end{aligned} \quad (3)$$

First order condition with respect to M which maximize the buyer's expected profit function i.e.

$$\frac{\partial E[TP_{b,M}]}{\partial M} = 0, \text{ gives the result}$$

$$M = \frac{\beta}{(e-\beta-1)} \left[c_b + \frac{A_b}{Q} + Ic_b E[\alpha]Q - c_s E[\alpha] \right], (\beta+1 < e) \quad (4)$$

Sufficient condition $\frac{\partial^2 E[TP_{b,M}]}{\partial M^2} < 0$ is also satisfied i.e. given Expected profit function of the buyer is concave with respect to M . Substituting Eq. (4) into Eq. (2), we have

$$p_b = \frac{e}{(e-\beta-1)(1-E[\alpha])} \left[c_b + \frac{A_b}{Q} + Ic_b E[\alpha]Q - c_s E[\alpha] \right], \beta+1 < e \quad (5)$$

Concavity of expected total profit $E[TP^c_b(p_b, M)]$ in Eq. (1) with respect to p_b and M is shown with the help of graph (Fig. 1)

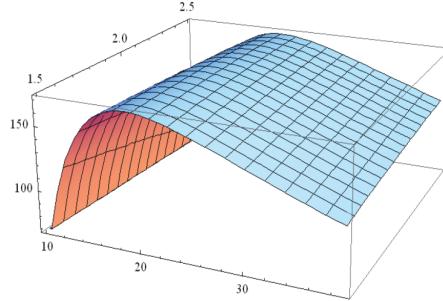


Fig. 1. Optimal buyer's total expected profit with respect to p_b and M

Since,

$$\frac{\partial^2 E[TP^c_b(p_b, M)]}{\partial p_b^2} > 0, \quad \frac{\partial^2 E[TP^c_b(p_b, M)]}{\partial M^2} > 0 \text{ and}$$

$$\left[\frac{\partial^2 E[TP^c_b(p_b, M)]}{\partial p_b^2} \right] \left[\frac{\partial^2 E[TP^c_b(p_b, M)]}{\partial M^2} \right] - \left[\frac{\partial^2 E[TP^c_b(p_b, M)]}{\partial p_b \partial M} \right]^2 < 0$$

3.2 Seller's model

The seller delivers a lot to the buyer. The aim of the seller is to find the optimal selling price, c_b and order quantity, Q to maximize his net profit. The sales revenue generated by seller is $c_b Q$, purchasing cost per year is cQ , ordering cost per year is A_s and holding cost of the seller is $\frac{1}{2} ICQT \frac{D}{r}$. The profit function of the seller is denoted by $TP_s(c_b, Q)$ and given by,

Seller's profit = Sales Revenue - Purchasing Cost - Ordering cost - Holding cost

$$TP_s(c_b, Q) = (c_b Q - CQ - A_s) - \frac{1}{2} ICQT_2 \frac{D}{r} \quad (6)$$

$$\text{Cycle length, } T_2 = \frac{Q}{D}$$

Expected profit of seller per cycle is,

$$E[TP^c_s(c_b, Q)] = \frac{D}{Q} (c_b Q - CQ - A_s) - \frac{1}{2} ICQu^{-1} = D(c_b - C - \frac{A_s}{Q}) - \frac{1}{2} ICQu^{-1}, \quad u = \frac{r}{D} \quad (7)$$

The aim is to calculate the optimal value of c_b and Q which maximize the seller's expected profit. The optimal value of Q can be obtained by differentiating Eq. (7) with respect to Q for fixed c_b and setting to zero, $\frac{\partial E[TP^c_s(c_b, Q)]}{\partial Q} = 0$, gives

$$Q = \sqrt{\frac{2A_s D}{ICu^{-1}}} \quad (8)$$

The expected profit of the seller $E[TP_s^c(c_b, Q)]$ function is concave in Q because second order condition is,

$$\frac{\partial^2 E[TP_s^c(c_b, Q)]}{\partial Q^2} = -2 \frac{A_s D}{Q^3} < 0 \quad (9)$$

Substituting the value of Q in the Eq. (7), we get

$$E[TP_s^c(c_b, Q)] = D \left[c_b - C - \frac{A_s}{\sqrt{\frac{2A_s D}{IC u^{-1}}}} \right] - \frac{1}{2} IC u^{-1} \sqrt{\frac{2A_s D}{IC u^{-1}}} \quad (10)$$

Solving $E[TP_s^c(c_b, Q)] = 0$ for zero profit gives,

$$c_{b_0} = C + \frac{A_s}{Q} + \frac{1}{2} IC(uD)^{-1} Q \quad (11)$$

Above profit function is increasing linear function in c_b , the optimal value can be obtained by setting the highest price value through negotiation by the seller to the buyer. Therefore,

$$c_b^* = F c_{b_0} = F \left(C + \frac{A_s}{Q} + \frac{1}{2} IC(uD)^{-1} Q \right) \text{ for some suitable, } F > 1 \quad (12)$$

3.3 Asymmetric/incomplete information game

Asymmetric information game is a game in which at least one player possesses some private information regarding the game but other players may not know that information. Just before making a move in the game, a player has some initial private information, called the type of the player. For example, in an auction theory considering a single individual item, each player would know a valuation for the items deterministically, while the other players may have only probabilistic information about how much this player value the items. Strategic game with incomplete information based on the works described in Harsanyi (1967, 1968) can be defined in seller buyer supply chain model as follows.

$N = \{P_1, P_2\}$ is the set of player. In our model, buyer and seller are two players which are represented by P_1 and P_2 . Since unit price is taken by the buyer and unit marketing expenditure influences the demand, $D = K p_b^{-e} M^\beta$ of the product and buyer has better knowledge about the demand but seller is not aware about the price elasticity, e and marketing expenditure, β which determine the market demand. In this model, all other input parameters are assumed to be known to both the players.

Let θ_1 be the type of buyer. Therefore, based on our assumption,

$$\theta_1 = \{(e, \beta); \beta = Le, 0 < \beta < 1\}$$

We denote a typical member of θ_1 by μ_1

$A = \{A_1, A_2\}$ is the set of actions of two player. These are also the decision variables of the players.

$$A_1 = \{p_b, M\} \text{ and } A_2 = \{C_b, Q\}.$$

Seller's unpredictability over the buyer's type can be expressed by the probability density function denoted by $g_1(\mu_1)$, $\mu_1 \in \theta_1$. The expected value is represented by E_{μ_1} .

4. The non-cooperative Stackelberg games

Non-Cooperative game theory is very useful in modelling and understanding multi economic problems characterized by strategic interdependency.

The non-cooperative Stackelberg strategic game structure was given in 1934 by Heinrich Stackelberg who published Market Structure and Equilibrium which illustrated the model and was a breaking point in the study of market structure. This is used to form the association between the partners of the supply-chain. In this model, two players, seller and buyer interact with each other. One player performs as the leader and take the first move initiative and another player acts as the follower, move sequentially and shows his best response based on given information. The aim of the leader is to create the best policies based on the best response given by the buyer to maximize his profit.

4.1 The seller-Stackelberg model with asymmetric information structure

In this model, buyer behaves as the follower while the seller behaves as the leader. Seller moves first and offers selling price, c_b , and order quantity, Q to the buyer in the situation under which demand is unknown to seller. The buyer chooses his best response according to the seller policy with the aim to increase his expected total profit. Based on the buyer's best move, the seller chooses his best strategy for finding the optimal selling price and lot size to gain his maximum profit.

$$\begin{aligned} \text{Max } E_{\mu_1} \text{TP}_s^c(c_b, Q) &= \int_{\mu_1} \left(\frac{D}{Q} (c_b Q - CQ - A_s) - \frac{1}{2} ICQ \frac{D}{r} \right) g_1(\mu_1) d\mu_1 \\ &= \int_{\mu_1} \left(\frac{1}{Q} K p_b^{-e} M^\beta (c_b Q - CQ - A_s) - \frac{1}{2} ICQu^{-1} \right) g_1(\mu_1) d\mu_1 \end{aligned} \quad (13)$$

subject to

$$p_b = \frac{e}{(e-\beta-1)(1-E[\alpha])} \left[c_b + \frac{A_b}{Q} + Ic_b E[\alpha]Q - c_s E[\alpha] \right], \beta + 1 < e \quad (14)$$

$$M = \frac{\beta}{(e-\beta-1)} \left[c_b + \frac{A_b}{Q} + Ic_b E[\alpha]Q - c_s E[\alpha] \right], \beta + 1 < e \quad (15)$$

Cycle length, $T = \max(T_1, T_2)$

Substituting the Eq. (14) and Eq. (15) into Eq. (13), the problem is converted into an unconstrained non-linear function with two variables c_b and Q , the optimal solution can be obtained by any software tool.

4.2 The seller-Stackelberg model with symmetric information structure

Under symmetric information structure, buyer and seller know about each other's activity carried out i.e. seller knows the market demand thus, element of randomness is ignored i.e. expectation operator can be omitted. Now, seller Stackelberg problem becomes

$$\begin{aligned} \text{Max } E_{\mu_1} (\text{TP}_s^c(c_b, Q)) &= \frac{D}{Q} (c_b Q - CQ - A_s) - \frac{1}{2} ICQ \frac{D}{r} \\ &= \frac{1}{Q} K p_b^{-e} M^\beta (c_b Q - CQ - A_s) - \frac{1}{2} ICQu^{-1} \end{aligned} \quad (16)$$

subject to

$$p_b = \frac{e}{(e - \beta - 1)(1 - E[\alpha])} \left[c_b + \frac{A_b}{Q} + Ic_b E[\alpha]Q - c_s E[\alpha] \right], \beta + 1 < e \quad (17)$$

$$M = \frac{\beta}{(e - \beta - 1)} \left[c_b + \frac{A_b}{Q} + Ic_b E[\alpha]Q - c_s E[\alpha] \right], \beta + 1 < e \quad (18)$$

Cycle length, $T = \max(T_1, T_2)$

Substituting the Eq. (14) and Eq. (15) into Eq. (13), the problem is converted into an unconstrained non-linear function with two variables c_b and Q , the optimal solution can be obtained by any software tool.

5. Numerical examples

This example shows the effect of the defective items in seller Stackelberg game model under asymmetric information structure. Input parameters in this example are taken from two papers Esmaeili and Zeephongsekul (2010) and Jaggi et al. (2013) which are given below. Suppose, $A_b = \$40$, $A_s = \$140$, $C = \$1.5$, $I = 0.1$, $u = 1.01$, $k = 3500$, $F = 1.25$, $L = 0.09$ and mean value $\bar{e} = 1.7$ standard deviation $\sigma_e = 0.01$. The fraction of imperfect quality item, α , is uniformly distributed on (a, b) , $0 < a < b < 1$ i.e., $\alpha \sim U(a, b)$. With $\alpha \sim U(a, b)$, $E[\alpha] = \frac{a+b}{2}$ and $E[(1-\alpha)^2]$ which can be determined with formula $E[(1-\alpha)^2] = \int_a^b (1-\alpha)^2 f(\alpha) d\alpha = \frac{a^2 + ab + b^2}{3} + 1 - a - b$. The expected value of α is $E[\alpha] = 0.02$, $E[(1-\alpha)^2] = 0.960$, where $a = 0$ and $b = 0.04$, $\lambda = 175200$ unit/year, $c_s = \$5$.

Example 1

An example is shown to show the effect of the defective items in the seller Stackelberg game model under symmetric information structure. Seller Stackelberg model gives the following optimal values, Eq. (16) gives the results, $Q^* = 249$ units and $c_b^* = \$5.92$. Eq. (17) and Eq. (18) yield the results, $p_b^* = \$18.88$, $M^* = \$1.63$. With these results, the seller's profit, $E[TP_{s^*}]^* = \$80.08$ and the buyer's expected profit, $E[TP_{b^*}]^* = \$211.11$

Example 2

An example is shown to show the effect of the defective items in the seller Stackelberg game model under asymmetric information structure. Seller Stackelberg model gives the following optimal values. Eq. (13) gives the results, $Q^* = 247$ units and $c_b^* = \$5.96$. Eq. (14) and Eq. (15) yield the results, $p_b^* = \$18.987$ and $M^* = \$1.63$, with these results, the seller's profit, $E[TP_{s^*}]^* = \$79.69$ and the buyer's expected profit, $E[TP_{b^*}]^* = \$210.37$.

6. Sensitivity analysis

Sensitivity analysis is performed to investigate the impact of parameters α , K on the different parameters c_b , p_b , M , D , Q , $E[TP_{s^*}]^*$, $E[TP_{b^*}]^*$ in the non-cooperative seller Stackelberg game model by taking both the information structure: symmetric and asymmetric. Results of sensitivity analysis are shown through the Tables (1-4).

Table 1

Seller Stackelberg's sensitivity analysis with respect to α in symmetric information structure

α^*	c_b^*	Q^*	p_b^*	M^*	D^*	$E[TP_{b^*}]^*$	$E[TP_{s^*}]^*$
0.005	6.10	244	19.37	1.70	24	206.05	80.75
0.010	6.04	246	19.21	1.68	25	207.52	80.52
0.015	5.982	247	19.044	1.655	25	209.46	80.30
0.020	5.92	249	18.88	1.63	25	211.11	80.08

0.025	5.87	250	18.707	1.61	26	213.08	79.86
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Table 2Seller Stackelberg's sensitivity analysis with respect to α in asymmetric information structure

α^*	c_b^*	Q^*	p_b^*	M^*	D^*	$E[TP^c_b]^*$	$E[TP^c_s]^*$
0.005	6.13	243	19.48	1.71	24	205.05	80.37
0.010	6.07	244	19.32	1.69	25	206.87	80.14
0.015	6.02	246	19.15	1.66	25	208.46	79.91
0.020	5.96	247	18.99	1.64	25	210.37	79.69
0.025	5.90	249	18.82	1.62	26	212.04	79.47

Table 3Seller Stackelberg's sensitivity analysis with respect to k in asymmetric information structure

k^*	c_b^*	Q^*	p_b^*	M^*	D^*	$E[TP^c_b]^*$	$E[TP^c_s]^*$
2500	6.46	195	20.72	1.79	16	130.57	52.27
3000	6.17	222	19.72	1.71	20	170.03	65.82
3500	5.96	247	18.99	1.64	25	210.33	79.69
4000	5.80	270	18.43	1.59	30	251.50	93.81
4500	5.67	292	17.98	1.56	35	293.14	108.15

Table 4Seller Stackelberg's sensitivity analysis with respect to k in symmetric information structure

k^*	c_b^*	Q^*	p_b^*	M^*	D^*	$E[TP^c_b]^*$	$E[TP^c_s]^*$
2500	6.42	197	20.58	1.78	16	131.01	52.52
3000	6.13	224	19.59	1.69	21	170.65	66.13
3500	5.92	249	18.88	1.63	26	211.11	80.08
4000	5.76	272	18.33	1.59	31	252.39	94.28
4500	5.63	294	17.88	1.55	36	294.18	108.69

Observations

In a symmetric information structure, it is reflected from the Table 1 that as the percentage of imperfect quantity items goes up, the optimal selling price of the buyer, selling price of the seller and marketing expenditure go down, but optimal order quantity increases. Since the demand and order quantity increase, the buyer's profit increases. In an asymmetric information structure, it is depicted from the Table 2 that as the percentage of imperfect quantity items increases, then the demand of the quantity and order quantity increase, which results in more profit to the buyer. We can also observe that the profit of both the players in asymmetric information structure is lesser as compared to symmetric information structure. Thus, it is advised for both the players that they should have cordial relation between them and information sharing environment for good business.

In both the symmetric and asymmetric information structure, it is reflected from the Tables 3 and 4 that when the value of K goes up, the selling price decreases marginally while the lot size increases and the present worth of profit of buyer and seller increases significantly. Further, it is also observed from Tables 3 and 4 that demand is increasing as k increases but profit is lesser in asymmetric information structure as compared to symmetric information structure.

7. Conclusions

This study considered supply chain model for imperfect quality items with asymmetric and symmetric information structure and interaction of two players is modeled by seller Stackelberg approach. In this paper, it is presumed that the demand function depends on the selling price of the buyer and marketing expenditure. Optimal solutions are obtained in this approach. Effect of defective items and scaling constant demand function on decision variables of both the players and their profits are shown through

sensitivity analysis in both the symmetric and asymmetric structure. Numerical examples show that in symmetric information structure, the buyer's and seller's profits are more than those obtained in the asymmetric structure. When demand function is not known to the seller, then seller's profit is lesser, even though seller is superior the buyer. Thus, both the players will be benefitted in case of complete and cordial information structure. This present model can be extended by taking the shortage cost, which is not considered in this paper. In this model, we have considered that defective items are sold at a discounted price. Instead of that, return policy can be applied, in this case transportation cost will be added.

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